```
In [581:
           #ElasticNet
            import pandas as pd
           import warnings
           warnings.filterwarnings("ignore")
           data=pd.read csv("/home/placement/Downloads/fiat500.csv")
In [59]:
           data.describe()
In [60]:
Out[60]:
                            ID engine power
                                              age in days
                                                                     km previous owners
                                                                                                   lat
                                                                                                               lon
                                                                                                                           price
                   1538.000000
                                              1538.000000
                                                             1538.000000
                                                                              1538.000000
                                                                                          1538.000000
                                                                                                       1538.000000
                                                                                                                    1538.000000
             count
                                  1538.000000
             mean
                     769.500000
                                    51.904421
                                              1650.980494
                                                            53396.011704
                                                                                 1.123537
                                                                                            43.541361
                                                                                                         11.563428
                                                                                                                    8576.003901
                                     3.988023
                                              1289.522278
                                                                                             2.133518
                                                                                                          2.328190
                                                                                                                    1939.958641
               std
                    444.126671
                                                            40046.830723
                                                                                 0.416423
                      1.000000
                                    51.000000
                                                366.000000
                                                             1232.000000
                                                                                 1.000000
                                                                                            36.855839
                                                                                                          7.245400
                                                                                                                    2500.000000
              min
              25%
                     385.250000
                                    51.000000
                                               670.000000
                                                            20006.250000
                                                                                 1.000000
                                                                                            41.802990
                                                                                                          9.505090
                                                                                                                    7122.500000
                     769.500000
                                    51.000000
                                                                                            44.394096
                                                                                                         11.869260
                                                                                                                    9000.000000
              50%
                                              1035.000000
                                                            39031.000000
                                                                                 1.000000
              75%
                   1153.750000
                                    51.000000
                                              2616.000000
                                                            79667.750000
                                                                                 1.000000
                                                                                            45.467960
                                                                                                         12.769040
                                                                                                                   10000.000000
              max 1538.000000
                                    77.000000
                                              4658.000000
                                                           235000.000000
                                                                                 4.000000
                                                                                            46.795612
                                                                                                         18.365520
                                                                                                                   11100.000000
In [61]:
           data.head()
Out[61]:
                   model engine_power age_in_days
                                                             previous owners
                                                                                     lat
                                                                                               Ion price
                                                       25000
             0
                1
                   lounge
                                     51
                                                 882
                                                                           1 44.907242
                                                                                          8.611560
                                                                                                   8900
                2
                      qoq
                                     51
                                                1186
                                                       32500
                                                                              45.666359 12.241890
                                                                                                   8800
             2
                     sport
                                     74
                                                4658
                                                     142228
                                                                              45.503300 11.417840
                                                                                                   4200
                   lounge
                                     51
                                                2739
                                                     160000
                                                                              40.633171 17.634609
                                                                                                   6000
                                     73
                                                     106880
                                                                           1 41.903221 12.495650
                      pop
In [62]: data1=data.loc[(data.previous owners)==1]
```

In [63]: data1

Out[63]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
0	1	lounge	51	882	25000	1	44.907242	8.611560	8900
1	2	рор	51	1186	32500	1	45.666359	12.241890	8800
2	3	sport	74	4658	142228	1	45.503300	11.417840	4200
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
4	5	рор	73	3074	106880	1	41.903221	12.495650	5700
1533	1534	sport	51	3712	115280	1	45.069679	7.704920	5200
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1535	1536	pop	51	2223	60457	1	45.481541	9.413480	7500
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990
1537	1538	pop	51	1766	54276	1	40.323410	17.568270	7900

1389 rows × 9 columns

In [64]: data2=data1.drop(['lat','lon','ID'],axis=1)

In [65]: data2

Out[65]:

	model	engine_power	age_in_days	km	previous_owners	price
0	lounge	51	882	25000	1	8900
1	pop	51	1186	32500	1	8800
2	sport	74	4658	142228	1	4200
3	lounge	51	2739	160000	1	6000
4	pop	73	3074	106880	1	5700
1533	sport	51	3712	115280	1	5200
1534	lounge	74	3835	112000	1	4600
1535	pop	51	2223	60457	1	7500
1536	lounge	51	2557	80750	1	5990
1537	pop	51	1766	54276	1	7900

1389 rows × 6 columns

In [66]: data2=pd.get\_dummies(data2)

In [67]: data2

$\sim$			
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	engine_power	age_in_days	km	previous_owners	price	model_lounge	model_pop	model_sport
0	51	882	25000	1	8900	1	0	0
1	51	1186	32500	1	8800	0	1	0
2	74	4658	142228	1	4200	0	0	1
3	51	2739	160000	1	6000	1	0	0
4	73	3074	106880	1	5700	0	1	0
1533	51	3712	115280	1	5200	0	0	1
1534	74	3835	112000	1	4600	1	0	0
1535	51	2223	60457	1	7500	0	1	0
1536	51	2557	80750	1	5990	1	0	0
1537	51	1766	54276	1	7900	0	1	0

1389 rows × 8 columns

```
In [68]: y=data2['price']
x=data2.drop('price',axis=1)
```

```
In [69]: y
Out[69]: 0
                   8900
                  8800
                  4200
          2
          3
                   6000
                   5700
          4
          1533
                  5200
          1534
                  4600
          1535
                  7500
          1536
                  5990
          1537
                  7900
          Name: price, Length: 1389, dtype: int64
In [90]: from sklearn.model_selection import train_test_split
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.33,random_state=42)
In [91]: x_test.head(5)
Out[91]:
                                         km previous_owners model_lounge model_pop model_sport
               engine_power age_in_days
                                 3347 148000
                                                                               0
           625
                        51
                                                         1
                                                                     1
                                                                                          0
           187
                        51
                                 4322
                                      117000
                                                         1
                                                                     1
                                                                               0
                                                                                           0
           279
                        51
                                 4322
                                      120000
                                                         1
                                                                     0
                                                                               1
                                                                                          0
           734
                        51
                                                                     0
                                  974
                                       12500
                                                         1
                                                                               1
                                                                                           0
           315
                                       37000
                        51
                                 1096
                                                         1
                                                                     1
                                                                               0
                                                                                          0
```

```
In [92]: y train
Out[92]: 915
                  10900
                   9700
          12
          638
                  10850
          190
                   9990
                  10300
          701
                   . . .
          1201
                   8300
                   3950
          1239
                   8900
          1432
          951
                   6500
          1235
                   8800
          Name: price, Length: 930, dtype: int64
In [93]: from sklearn.linear model import ElasticNet
          from sklearn.model selection import GridSearchCV
          elastic = ElasticNet()
          parameters = {'alpha': [1e-15, 1e-10, 1e-8, 1e-4, 1e-3,1e-2, 1, 5, 10, 20]}
          elastic regressor = GridSearchCV(elastic, parameters)
          elastic regressor.fit(x train, y train)
Out[93]: GridSearchCV(estimator=ElasticNet(),
                        param grid={'alpha': [1e-15, 1e-10, 1e-08, 0.0001, 0.001, 0.01, 1,
                                               5, 10, 201})
          In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
          On GitHub, the HTML representation is unable to render, please try loading this page with nbyiewer.org.
In [94]: elastic regressor.best params
Out[94]: {'alpha': 0.01}
```

localhost:8888/notebooks/ElasticNet.ipynb

```
In [95]: elastic=ElasticNet(alpha=.01)
    elastic.fit(x_train,y_train)
    y_pred_elastic=elastic.predict(x_test)

In [96]: from sklearn.metrics import r2_score
    r2_score(y_test,y_pred_elastic)

Out[96]: 0.8602162350730707

In [97]: from sklearn.metrics import mean_squared_error
    elastic_Error=mean_squared_error(y_pred_elastic,y_test)
    elastic_Error
Out[97]: 515349.9787871871
```

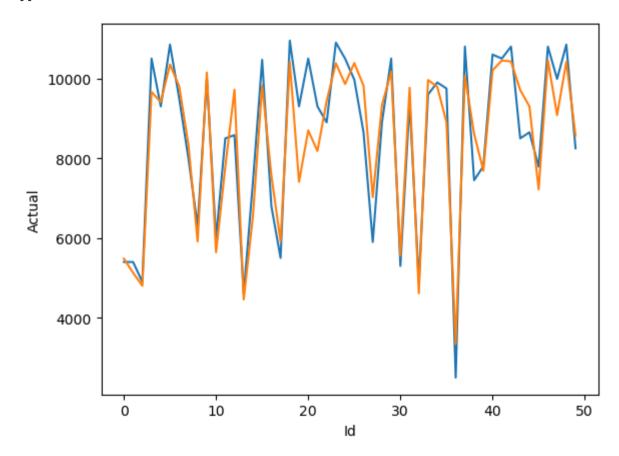
In [99]: Results=pd.DataFrame(columns=['Actual','Predicted'])
 Results['Actual']=y\_test
 Results['Predicted']=y\_pred\_elastic
 Results=Results.reset\_index()
 Results['Id']=Results.index
 Results.head(15)

## Out[99]:

	index	Actual	Predicted	ld
0	625	5400	5482.171479	0
1	187	5399	5127.531740	1
2	279	4900	4803.203231	2
3	734	10500	9662.825235	3
4	315	9300	9408.645424	4
5	652	10850	10350.952605	5
6	1472	9500	9806.127960	6
7	619	7999	8341.142824	7
8	992	6300	5913.786719	8
9	1154	10000	10149.093829	9
10	757	6000	5643.649619	10
11	1299	8500	7780.541311	11
12	400	8580	9720.293317	12
13	314	4600	4459.155236	13
14	72	7400	6541.667411	14

```
In [100]: import seaborn as sns
import matplotlib.pyplot as plt
sns.lineplot(x='Id',y='Actual',data=Results.head(50))
sns.lineplot(x='Id',y='Predicted',data=Results.head(50))
plt.plot()
```

## Out[100]: []



In [ ]:	